

Simplifying Generative AI App Development: Why Standards Matter

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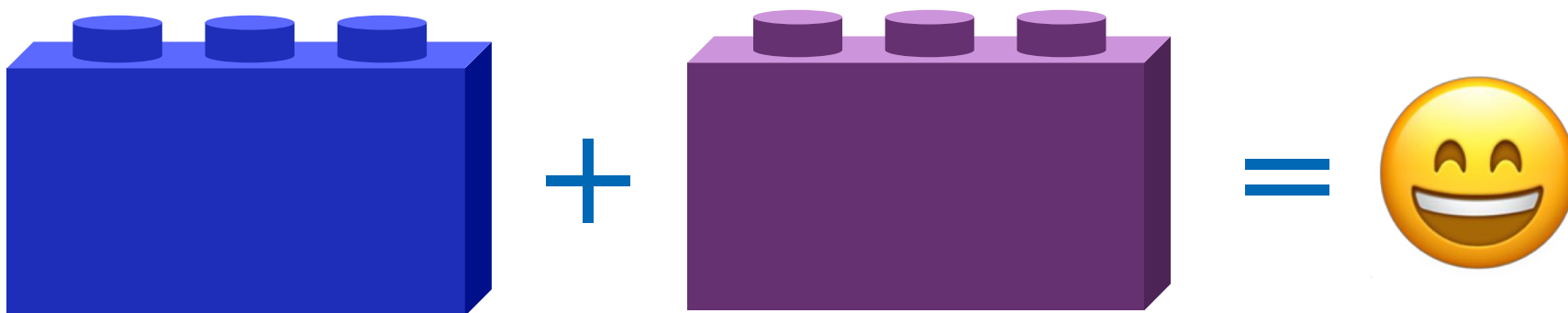
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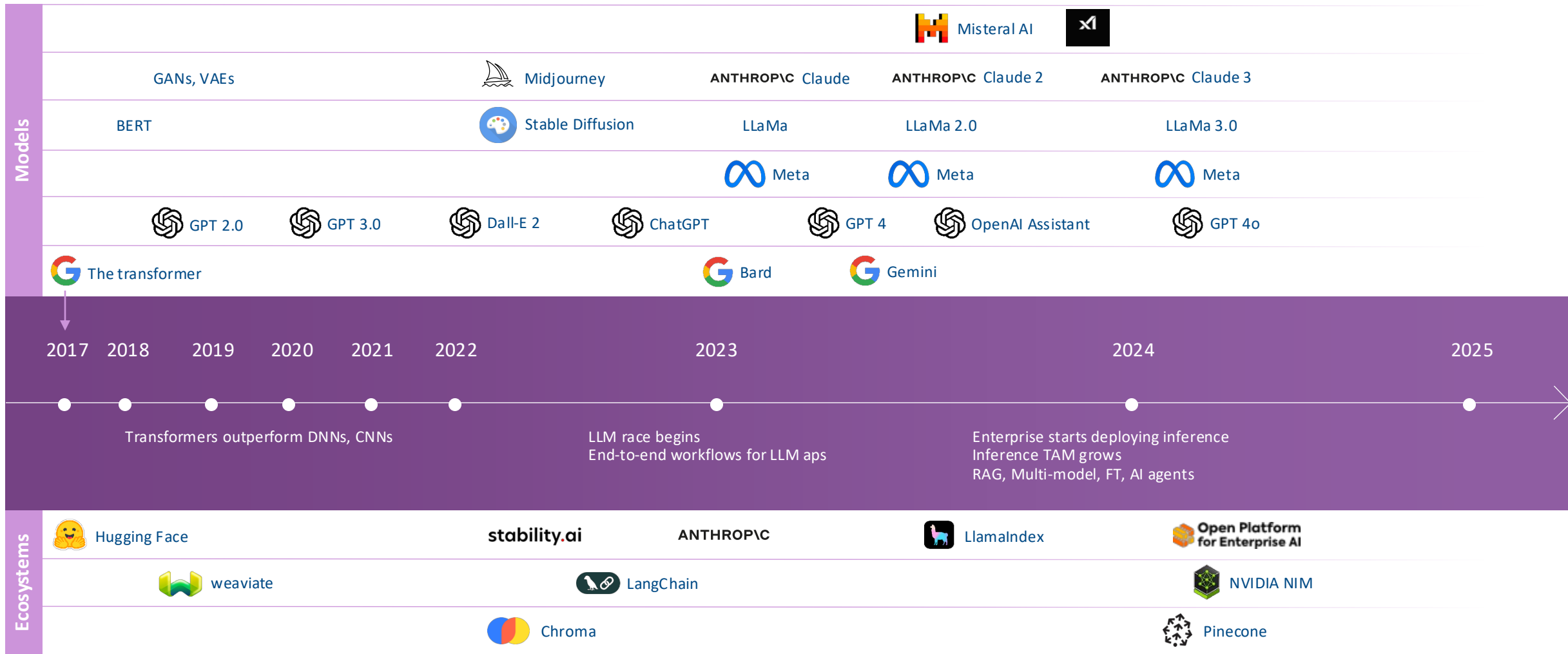
GenAI is emerging rapidly, but enterprises are struggling to realize GenAI value in production.

Why standardization matters

Life is easier when things go together



GenAI: transformers to AI Agents in 8 years



Massive innovation, expensive duplication, no standardization

- So many options, how do decide?
- Constantly reinventing the wheel
- Few best practices, wasted dev cycles.
- Multiple, circuitous routes to the same place
- Can we be pioneers without so much chaos?



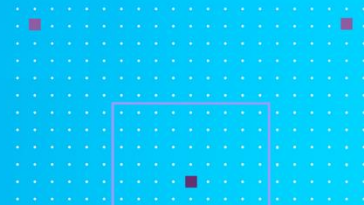
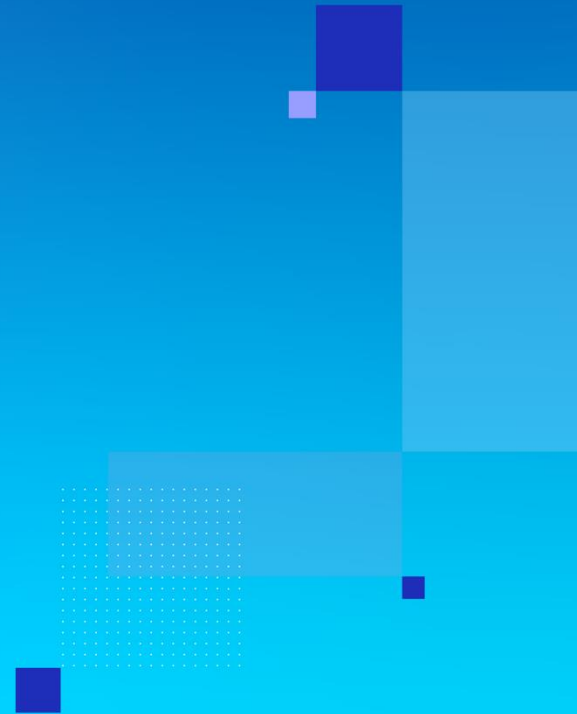
We've created a monster!

Without standards Gen AI RAG can be a beast!



The solution

A platform that tames the chaos



Open source platform that organizes GEN AI chaos



**Open Platform
for Enterprise AI**

- Composable building blocks for generative AI systems
- Integrates LLMs, data stores, and prompt engines
RAG AI blueprint with component stack structure and end-to-end workflow
- Additional stacks for translation, code generation, images
- Generative AI assessment tool for performance, features, trustworthiness and enterprise-grade readiness

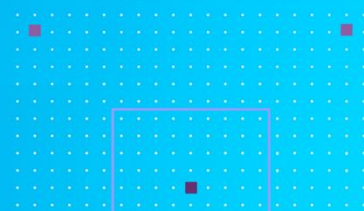
OPEA Contributors



Open Platform
for Enterprise AI

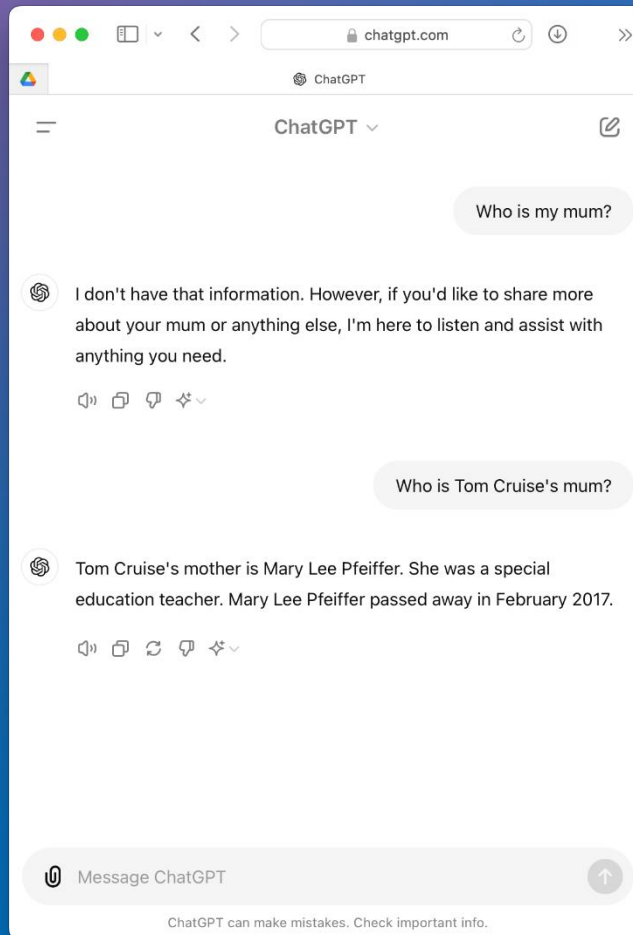
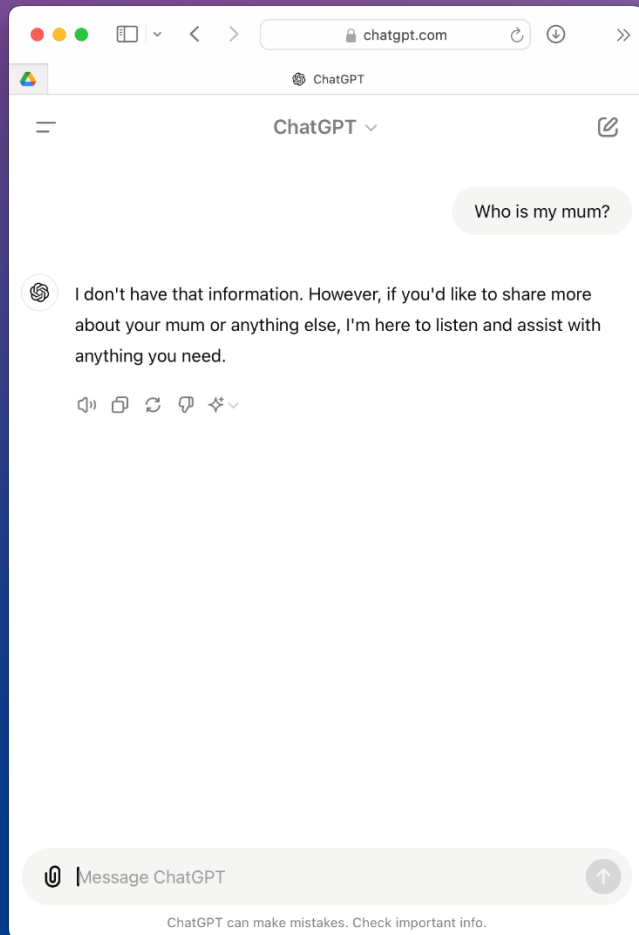
				东方国信	
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					by deepset
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	by Broadcom				

What is RAG?



What is RAG

LLMs don't know everything ...



But they do know a lot about
their training data

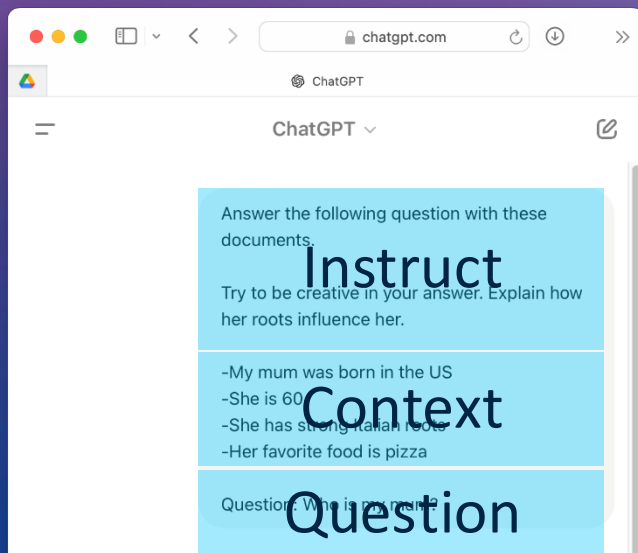
YOU TOLD ME YOU'RE SMART!



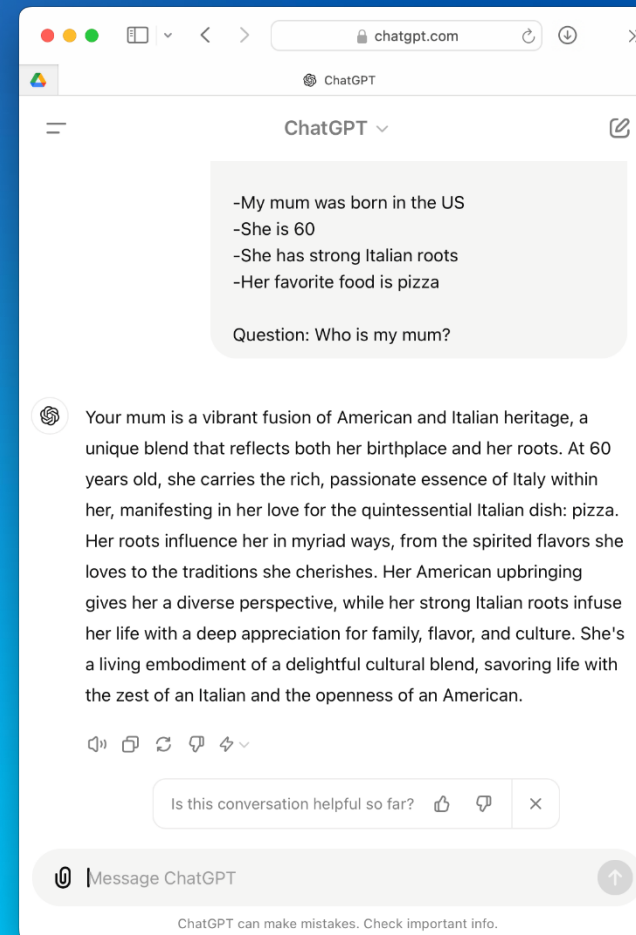
I'm a pre-trained model with general knowledge. You should have given me context.

What is RAG

If we augment an LLM with new context ...



It can generate a response to a novel topic



RETRIEVAL



Retrieve data from a knowledge base
(my brain in this case!)

AUGMENTED



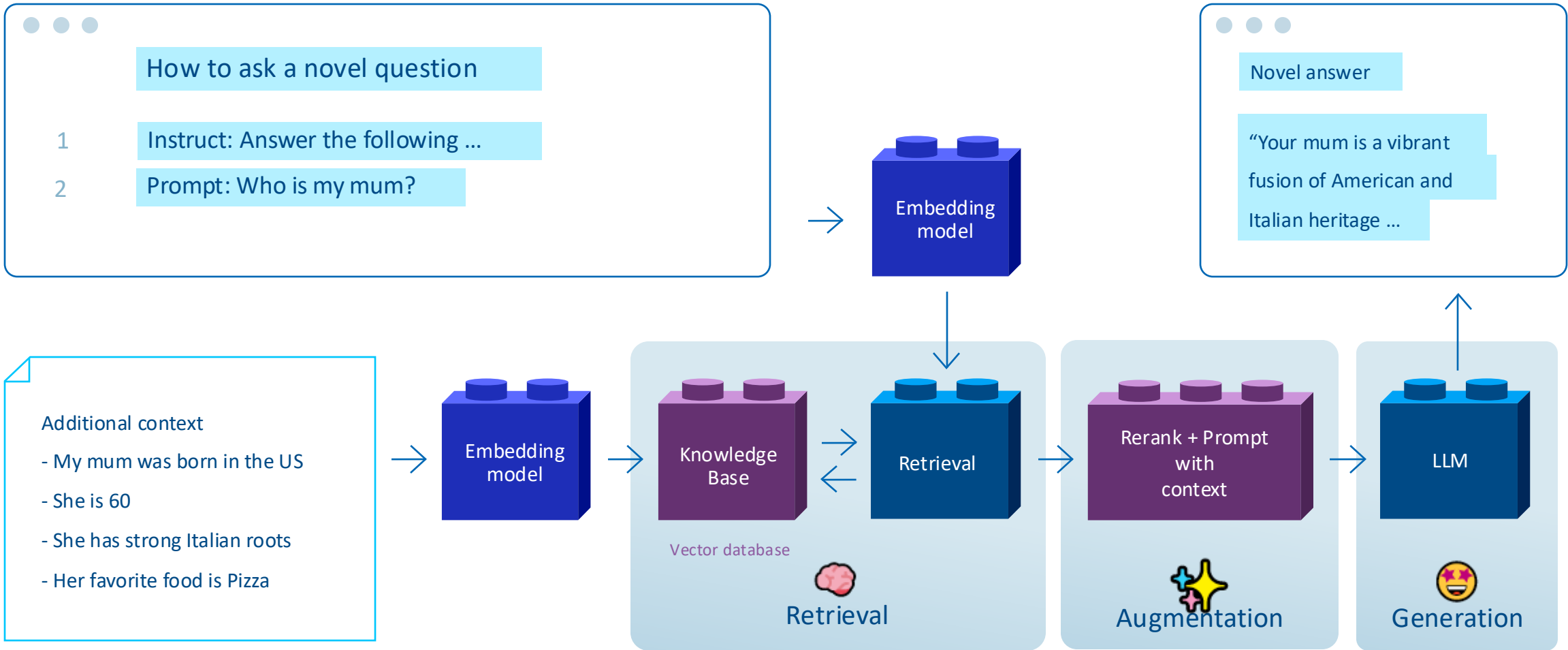
Give the model instructions
and context with the prompt

GENERATION



Get a response based on the retrieved
data

RAG process flow



Lack of standards make Gen AI extremely complex

Ecosystem complexity for Gen AI

Data

- Indexing the data
- Creating embeddings (vector representations)
- Choosing storage (ISVs?)
- Preparing the data

LLM

- Selecting the right model
- Interacting with multiple models
- Choosing frameworks

Deploy

- Hardware efficiency
- Retrieval method
- Applying guardrails
- Retrieval accuracy (reranking)
- Putting it all together
- Finding/building a megaservice

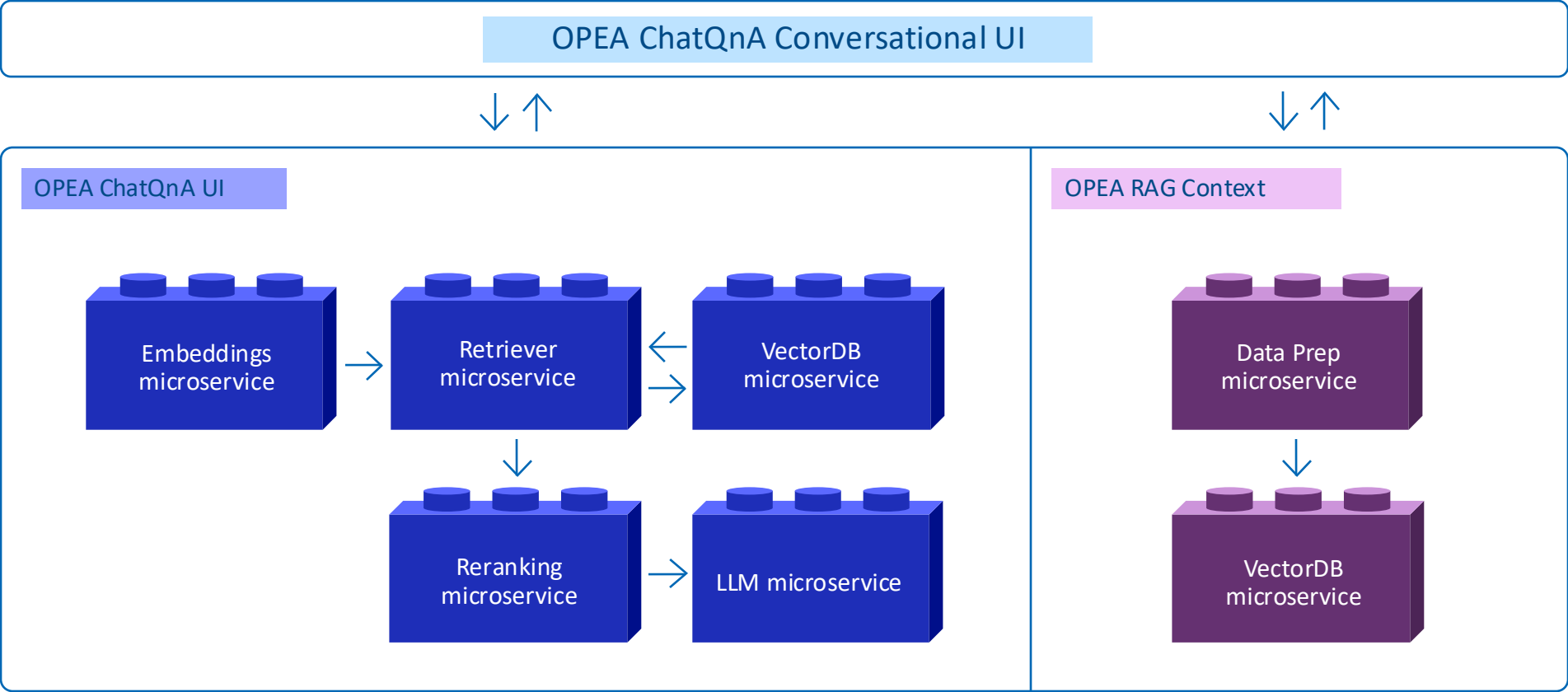
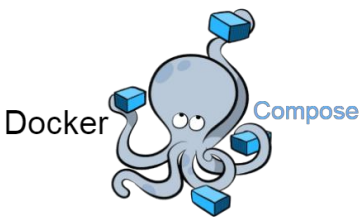
OPEA RAG reference meets these challenges

- Composable microservices for data, LLMs, and deployment
- Plug and play with 3rd party services—point to whichever data providers, embedding engines, and models you choose
- Over-arching mega service in place for faster development



DEMO

Under the hood – OPEA ChatQnA with RAG



- Chat with RAG running on microservice architecture
- Microservices can point to choice of services/APIs

Embeddings microservice - Converts from text to embeddings (vectors)



opea/embedding-tei ☆0

By [opea](#) · Updated 3 days ago

IMAGE

Variables

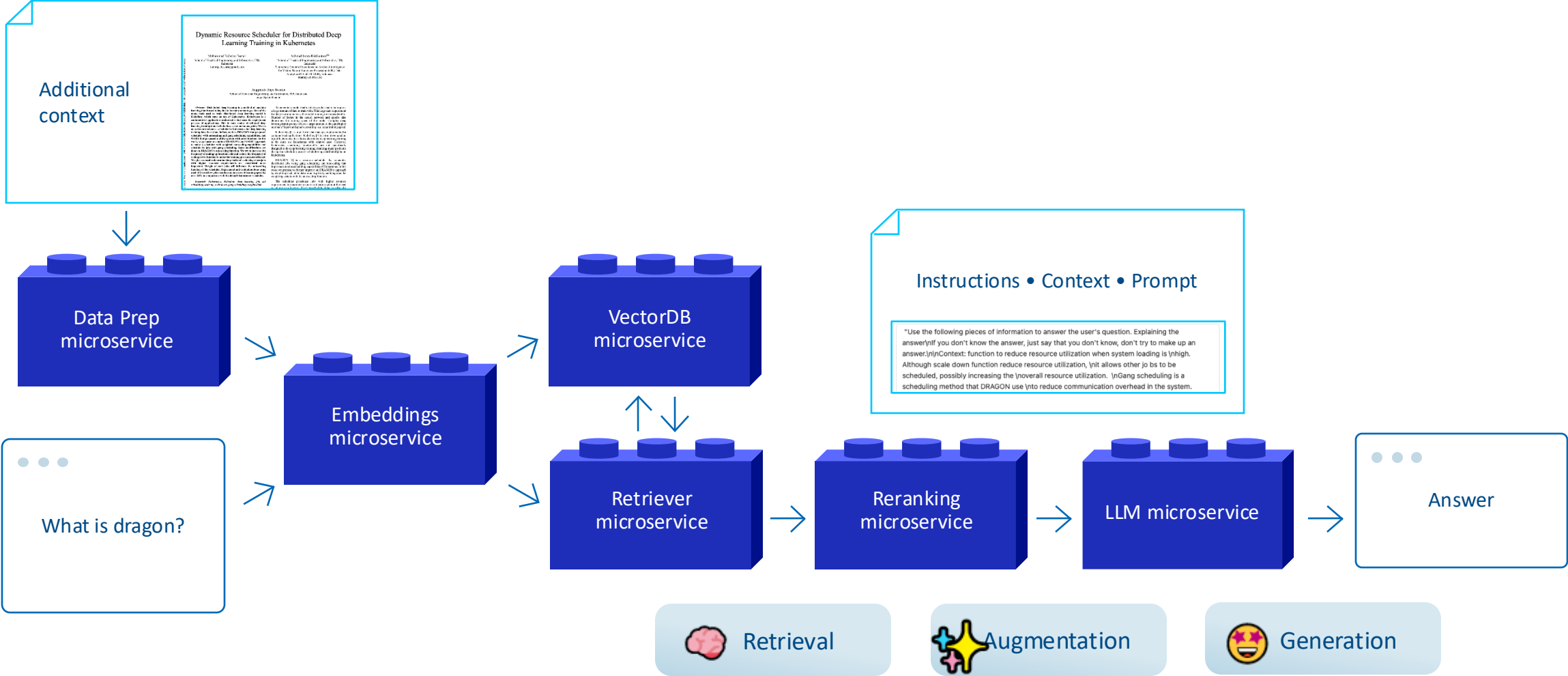
```
services:
  embedding:
    image: opea/embedding-tei:latest
    container_name: embedding-tei-server
    ports:
      - "6000:6000"
    ipc: host
    environment:
      http_proxy: ${http_proxy}
      https_proxy: ${https_proxy}
      TEI_EMBEDDING_ENDPOINT: ${TEI_EMBEDDING_ENDPOINT}
      LANGCHAIN_API_KEY: ${LANGCHAIN_API_KEY}
    restart: unless-stopped
```

Vector for “Who is my mum?”

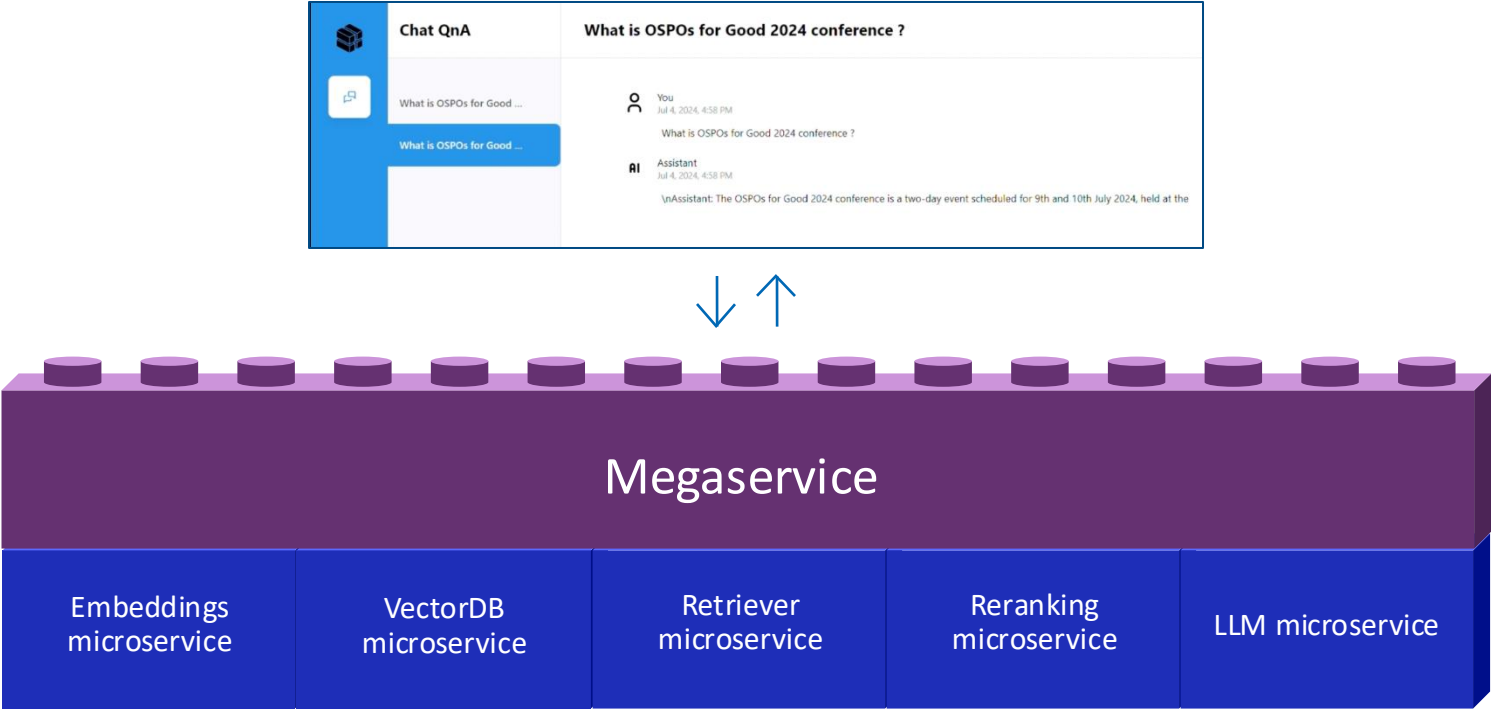
```
curl localhost:$your_port/embed -X POST -d  
'{"inputs": "Who is my mum?"}' -H 'Content-Type:  
application/json'
```

```
[[0.018129222,0.0030305043,-0.049874727,-  
0.035030127,0.014229514,-0.023594731.....-  
0.033771276,-0.0009737879,-0.006777766,-  
0.058805678,0.011158894,0.012094927,0.01805739,  
0.054448325,-0.032204594,0.049175356]]
```

How OPEA ChatQnA answers on context



OPEA ChatQnA Megaservice



CODE

Check out ChatQnA on GitHub



RAG conclusions and challenges

The good

- RAG is a great approach to having context-based answers
- OPEA makes the building blocks easy to deploy

The challenges

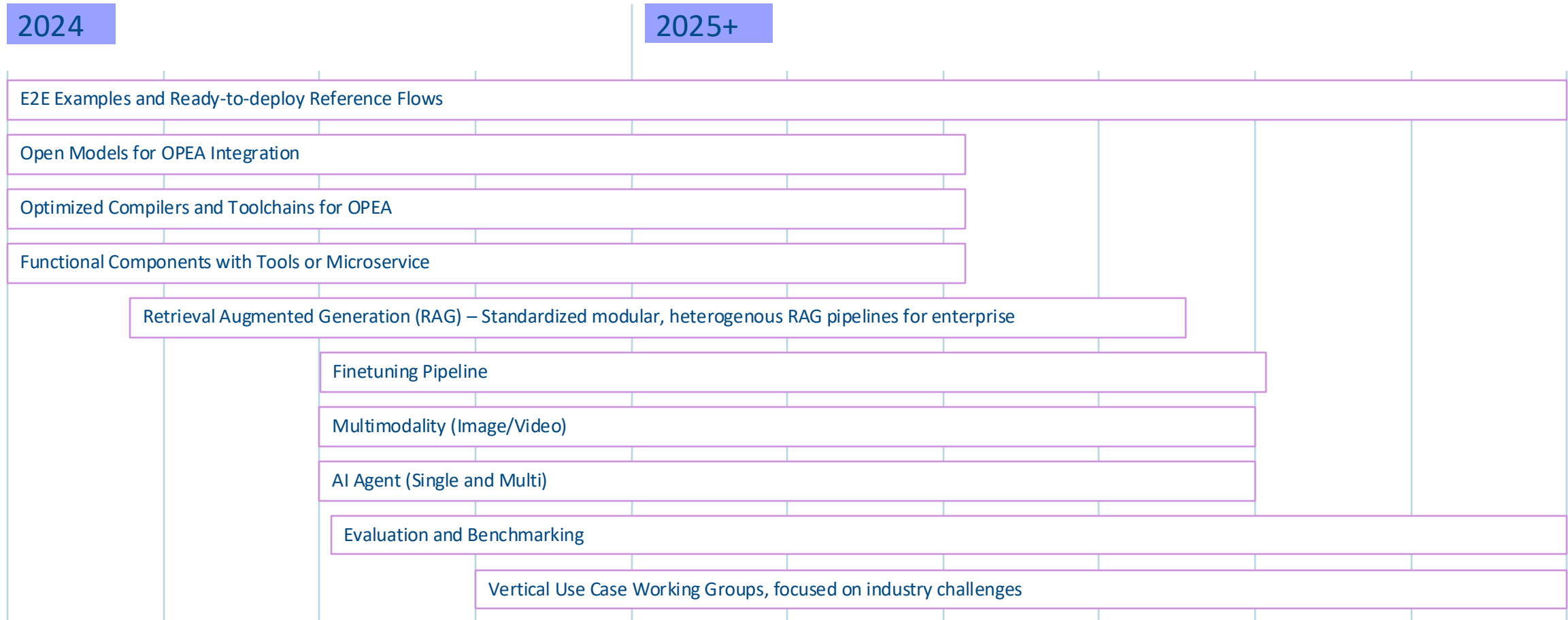
- Dealing with a large amount of data (Play with different similar search techniques)
- Different types of data (Images/Tables) – Intel Tech on Medium articles
- Is it all about indexing? Explore Retrieval augmented text-to-SQL generation



Join us!

There is more—a lot
more—you can build!

OPEA* Roadmap



*Preliminary

Help build a better OPEA, contribute!

opea-project / GenAICompsPublic

NotificationsFork 66Star 28

<> CodeIssues 45Pull requests 23ActionsProjects 1SecurityInsights

Prediction Guard Components (Running on Gaudi in ITDC) #343

New issue

Closed

dwhitena wants to merge 72 commits into opea-project:main from predictionguard:main

Conversation 0

Commits 72

Checks 13

Files changed 86

+1,543-254

dwhitena commented last week

Description

This PR does the following:

1. Integrates Prediction Guard powered OPEA components for LLM, LVM, embeddings, and guardrails. Prediction Guard hosts open access models on top of Gaudi in Intel Tiber Developer Cloud (ITDC). Users of Prediction Guard can use the models via a multi-tenant system (and an API key) or single-tenant/self-hosted, which provides a pathway for OPEA users to have turn-key model hosting of open models on top of Gaudi. That is, OPEA users can utilize OPEA to spin up OPEA compliant components with models hosted in Prediction Guard, which unlocks the various tooling and examples that will be built in an OPEA compliant ways.

2. Organizes the Guardrails component directory to allow for multiple providers of guardrails of various types. Rather than having a single way to run a single type of guardrail. The directory is now organized into directories for each type of guardrail (factuality, pii, etc.) with compatible components (from potentially different providers) underneath.

3. Organizes the LVM component directory to allow for multiple providers of LVMs. Currently there is the option to run the LVM component via transformers and Prediction Guard.

4. Fixes various typos or minor README clarifications as they were found.

We wanted to pull in all of these together to make sure that they all can work together as OPEA compliant components in the GenAIEamples. We tested each service individually and orchestrated together.

Issues

Reviewers

hshen14

kding1

ftian1

Assignees

No one assigned

Labels

None yet

Projects

None yet

Milestone

No milestone

Development

Successfully merging this pull request may close these issues.

Participate in OPEA



[Visit OPEA.dev](https://opea.dev)

- Join a Working Group
- Bring Enterprise AI use cases
- Contribute code, docs, projects, blueprints, and more
- Provide feedback
- Evangelism and Promotion of OPEA in YOUR communities, events

Scan for tools, sources, and resources

Visit the Intel Open Ecosystem Community and Evangelism GitHub page

- A PDF of this presentation
- Guides and community resources
- Links to articles and source material





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Open Platform for Enterprise AI

Thank you!



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